

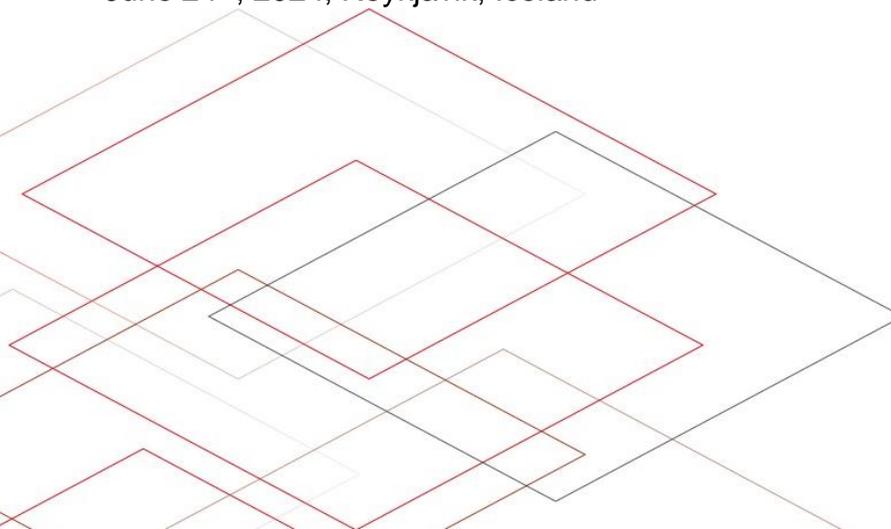
# Introduction: Open Research Knowledge Graph

**Oliver Karras, Alessio Ferrari, Davide Fucci, and Davide Dell'Anna**

[oliver.karras@tib.eu](mailto:oliver.karras@tib.eu), [alessio.ferrari@isti.cnr.it](mailto:alessio.ferrari@isti.cnr.it), [davide.fucci@bth.se](mailto:davide.fucci@bth.se), [d.dellanna@uu.nl](mailto:d.dellanna@uu.nl)

32nd IEEE International Requirements Engineering 2024 Conference – Exploring New Horizons: Expanding the Frontiers of Requirements Engineering

June 24<sup>th</sup>, 2024, Reykjavik, Iceland

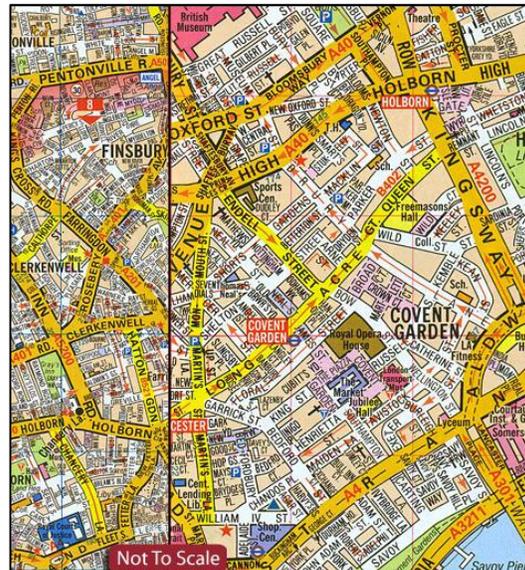


# Once Upon a Time, we Communicated with Paper...

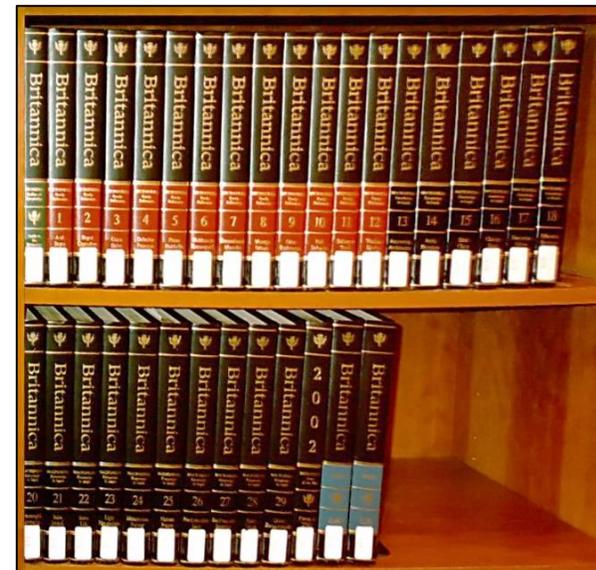
Who still remembers?



Mail order catalogs



Maps

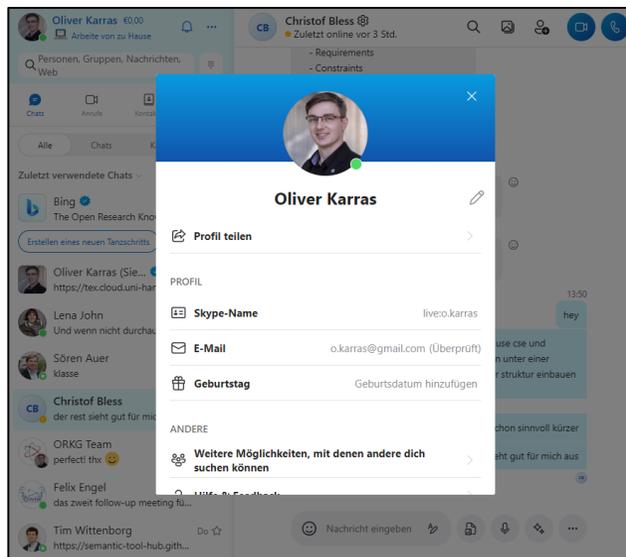
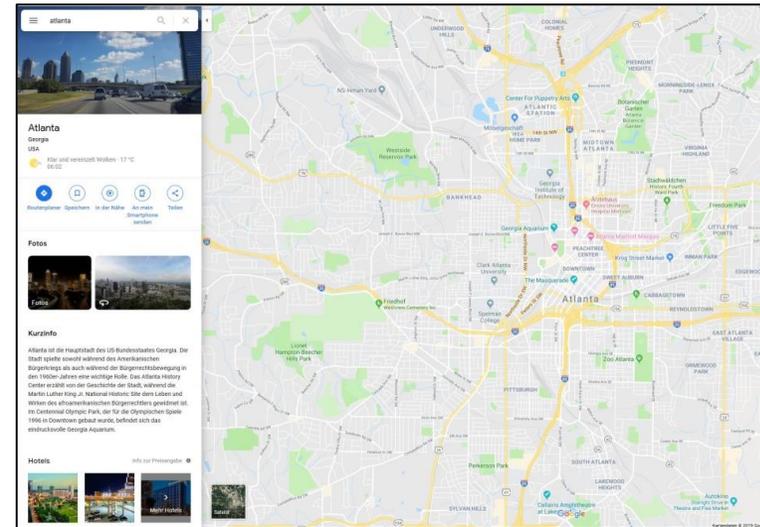
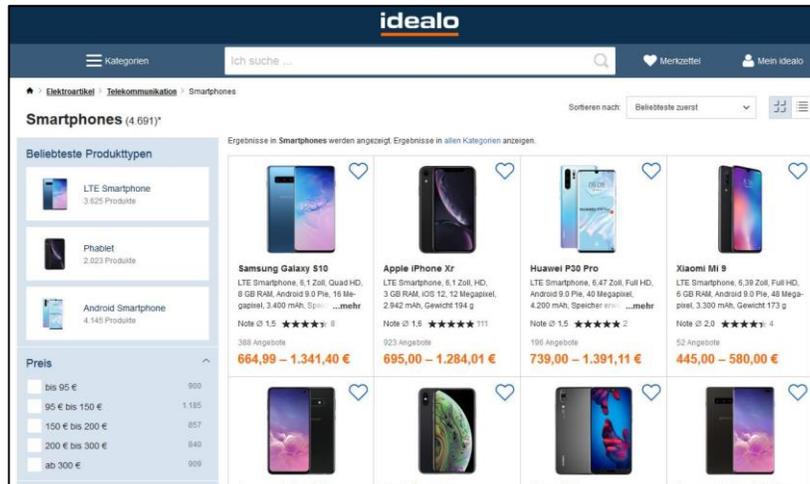


Encyclopedia



Phone books

# ... until Digital Transformation (Digitalization)!



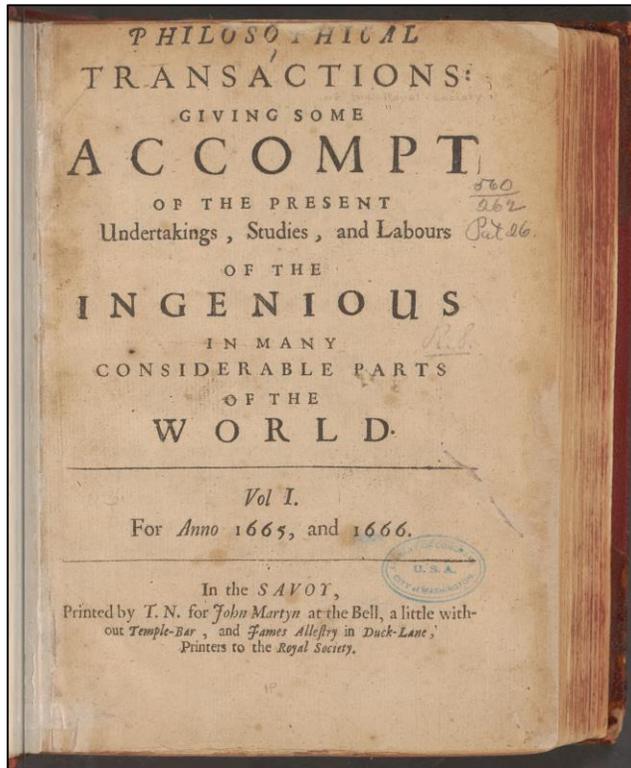
## The World of Publishing & Communication has Profoundly Changed!

- **New means adapted to the new possibilities**, e.g., platforms
- **Completely new business models**
- **More focus on data, interlinking, services, and search**
- **Integration, crowdsourcing, and data curation** are important

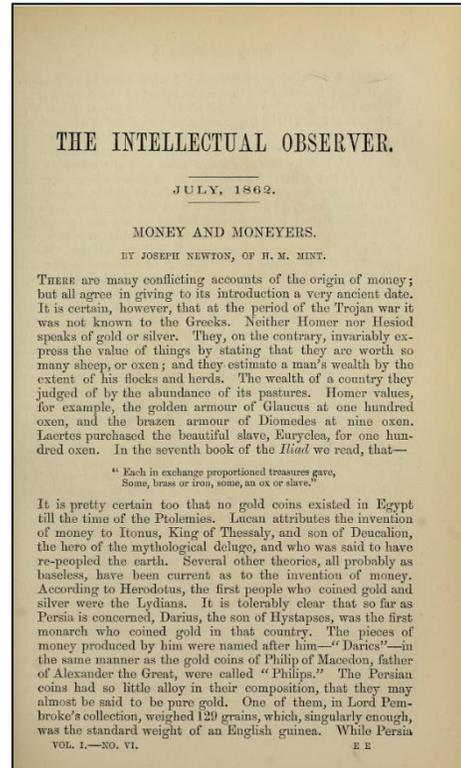
# **What has happened in academia in terms of scholarly publishing & communication?**

# Let's Take a Look

## 17<sup>th</sup> century



## 19<sup>th</sup> century



## 20<sup>th</sup> century

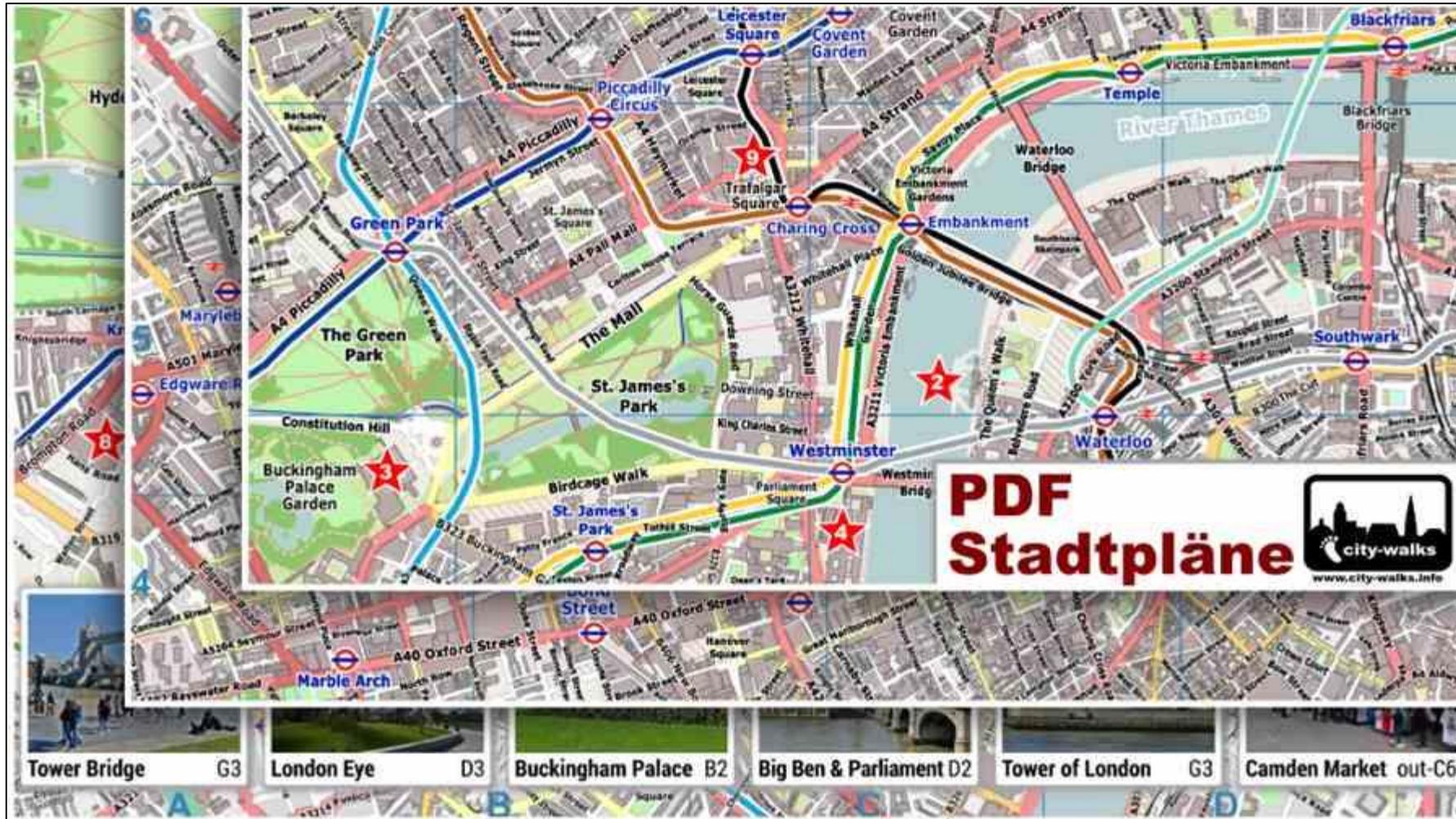


## 21<sup>st</sup> century



Scholarly publishing & communication has **not** changed (much)!

# Let's Take a Look



21<sup>st</sup> century

## AGDISTIS - Graph-Based Disambiguation of Named Entities Using Linked Data

Ricardo Usbeck<sup>1,2</sup>, Axel-Cyrille Ngonga Ngomo<sup>1</sup>, Michael Röder<sup>1,2</sup>, Daniel Gerber<sup>1</sup>, Sandro Athaide Coelho<sup>3</sup>, Sören Auer<sup>4</sup>, and Andreas Both<sup>2</sup>

<sup>1</sup> University of Leipzig, Germany  
<sup>2</sup> R&D, Unister GmbH, Germany  
<sup>3</sup> Federal University of Juiz de Fora, Brazil  
<sup>4</sup> University of Bonn & Fraunhofer IAIS, Germany  
(usbeck,ngonga}@informatik.uni-leipzig.de

**Abstract.** Over the last decades, several billion Web pages have been made available on the Web. The ongoing transition from the current Web of unstructured data to the Web of Data yet requires scalable and accurate approaches for the extraction of structured data in RDF (Resource Description Framework) from these websites. One of the key steps towards extracting RDF from text is the disambiguation of named entities. While several approaches aim to tackle this problem, they still achieve poor accuracy. We address this drawback by presenting AGDISTIS, a novel knowledge-base-agnostic approach for named entity disambiguation. Our approach combines the Hypertext-Induced Topic Search (HITS) algorithm with label expansion strategies and string similarity measures. Based on this combination, AGDISTIS can efficiently detect the correct URIs for a given set of named entities within an input text. We evaluate our approach on eight different datasets against state-of-the-art named entity disambiguation frameworks. Our results indicate that we outperform the state-of-the-art approach by up to 2% F-measure.

### 1 Introduction

The vision behind the Web of Data is to provide a new machine-readable layer to the Web where the content of Web pages is annotated with structured data (e.g., RDEa [1]). However, the Web in its current form is made up of at least 15 billion Web pages.<sup>1</sup> Most of these websites are unstructured in nature. Realizing the vision of a usable and up-to-date Web of Data thus requires scalable and accurate natural-language-processing approaches that allow extracting RDF from such unstructured data. Three tasks play a central role when extracting RDF from unstructured data: named entity recognition (NER), named entity disambiguation (NED), also known as entity linking [16], and relation extraction (RE). For the first sentence of Example 1, an accurate named entity recognition approach would return the strings Barack Obama and Washington,

<sup>1</sup> Data gathered from <http://www.worldwidewebsize.com/> on January 4th, 2014.

# Let's Take a Look



# Rethink How Scientific Knowledge is Communicated



*“The lightbulb was **not** invented by improving the candle.”*

**Oren Harari**

Digitalization is **more** than just Digitization!

Current and future scientific challenges can not be tackled with an outdated communication system.

---

**Digitalize Knowledge,  
Not Documents!**

# Example: Requirements Engineering and Empirical Research

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🔍

📄
Artikel

Ungefähr 3.970.000 Ergebnisse (0,16 Sek.)

Beliebige Zeit

Seit 2024

Seit 2023

Seit 2020

Zeitraum wählen...

---

Nach Relevanz sortieren

Nach Datum sortieren

---

Beliebige Sprache

Seiten auf Deutsch

---

Alle Typen

Übersichtsarbeiten

---

Patente einschließen

Zitate einschließen

[HTML] Empirical research in requirements engineering: trends and opportunities

T Ambreen, N Ikram, M Usman, M Niazi - Requirements Engineering, 2018 - Springer

... trends and future **research** directions. To represent a state-of-the-art of **requirements engineering**, along with various trends and opportunities of **empirical RE research**, we conducted a ...

☆ Speichern
🔗 Zitieren
Zitiert von: 125
Ähnliche Artikel
Alle 7 Versionen

[HTML] Empirical research methodologies and studies in Requirements Engineering: How far did we come?

M Daneva, D Damian, A Marchetto, O Pastor - Journal of systems and ..., 2014 - Elsevier

... Software **Engineering** paradigm. We summarize prior **empirical research** in RE and introduce the contributors to this special issue on **empirical research** methodologies and studies in ...

☆ Speichern
🔗 Zitieren
Zitiert von: 67
Ähnliche Artikel
Alle 12 Versionen

A systematic literature review of empirical research on quality requirements

T Olsson, S Sentilles, E Papatheocharous - Requirements Engineering, 2022 - Springer

... We want to understand the **empirical research** on quality **requirements** topics as well as ... We conclude that more **research** is needed as **empirical research** on quality **requirements** is not ...

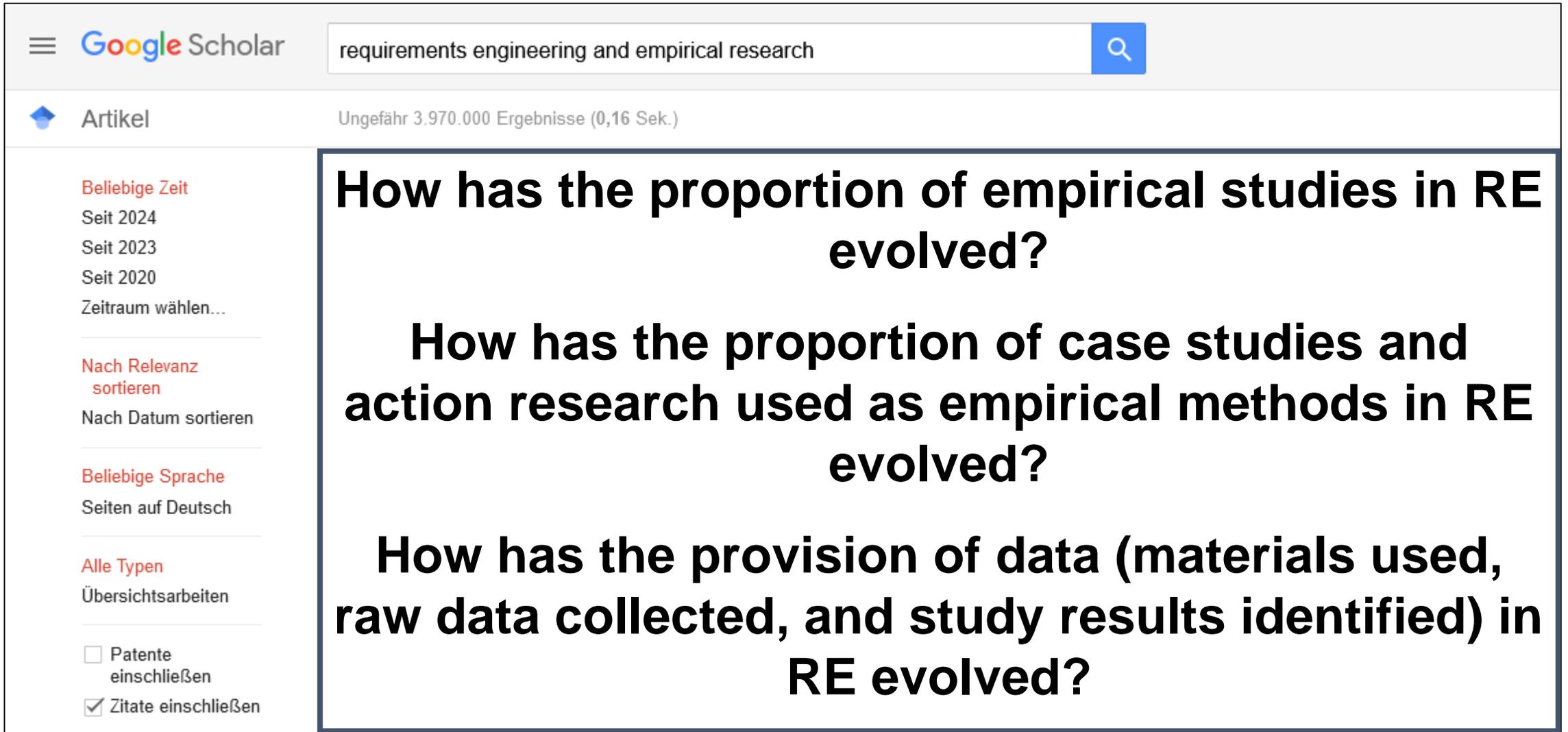
☆ Speichern
🔗 Zitieren
Zitiert von: 13
Ähnliche Artikel
Alle 9 Versionen

[HTML] springer.com

[HTML] sciencedirect.com

[PDF] springer.com

# Example: Requirements Engineering and Empirical Research



The image shows a screenshot of a Google Scholar search results page. The search query is "requirements engineering and empirical research". The results are filtered to "Artikel" (Articles) and show approximately 3,970,000 results in 0.16 seconds. The left sidebar contains filters for time period, relevance, language, and document type. The main content area is highlighted with a blue border and contains three research questions.

Google Scholar

requirements engineering and empirical research

Artikel

Ungefähr 3.970.000 Ergebnisse (0,16 Sek.)

**Beliebige Zeit**  
Seit 2024  
Seit 2023  
Seit 2020  
Zeitraum wählen...

**Nach Relevanz sortieren**  
Nach Datum sortieren

**Beliebige Sprache**  
Seiten auf Deutsch

**Alle Typen**  
Übersichtsarbeiten

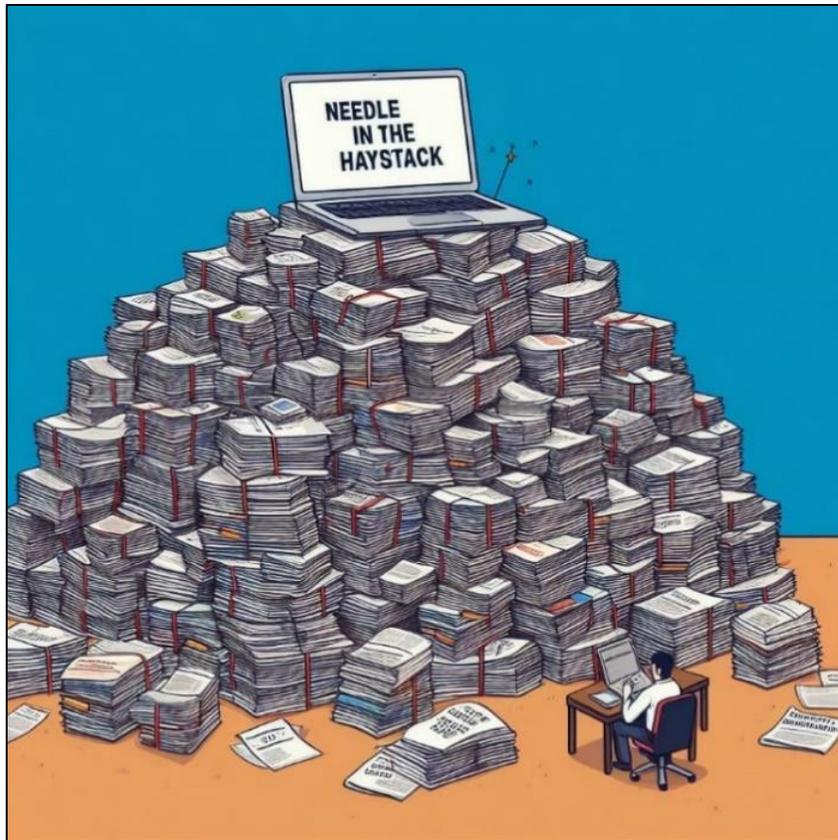
Patente einschließen  
 Zitate einschließen

**How has the proportion of empirical studies in RE evolved?**

**How has the proportion of case studies and action research used as empirical methods in RE evolved?**

**How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved?**

# How do We Answer These Questions so far?

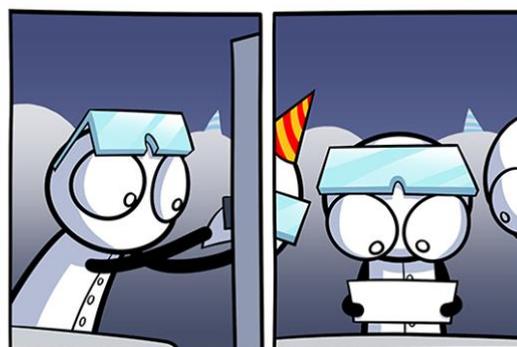
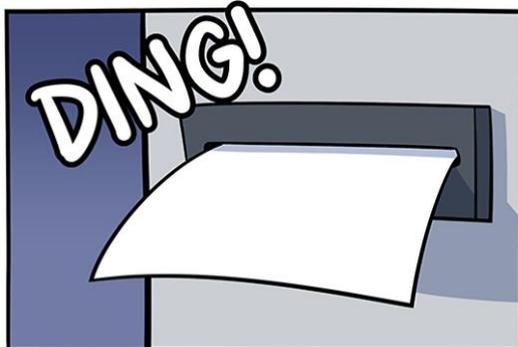
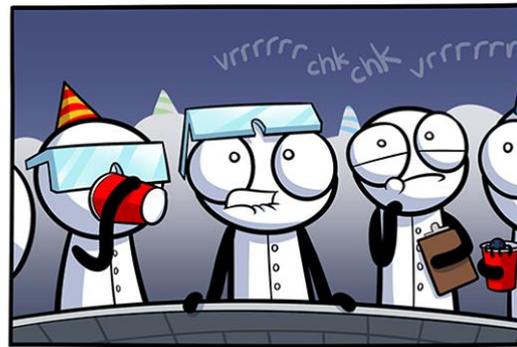
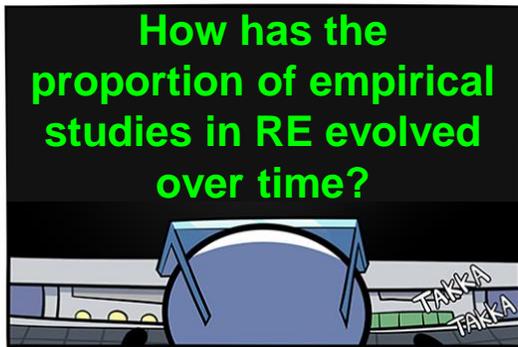
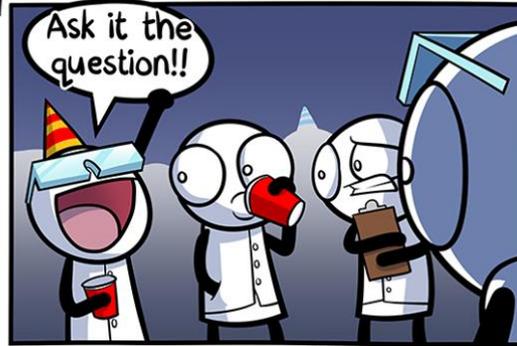


(S)LRs and SMSs for a **comprehensive, up-to-date, and long-term available** overview.



Over **7 million publications** per year<sup>[4]</sup> with an **increasing** (exponential) tendency.

# Wouldn't it be Great if we Could Ask the Computer?



How can we achieve this goal?

# Open Research Knowledge Graph (ORKG)

The screenshot shows the ORKG website interface. At the top, there is a navigation bar with the ORKG logo, menu items (View, Tools, About), a dropdown for 'NFDI4DataScience', a search bar, and buttons for '+ Add new' and 'Sign in'. Below the navigation is a hero section with the text 'Scholarly Knowledge. FAIR.' and a sub-header: 'The Open Research Knowledge Graph (ORKG) aims to describe research papers in a structured manner. With the ORKG, papers are easier to find and compare. Play video'. The main content area is titled 'Browse by research field' and features five red buttons representing different fields: Arts and Humanities (432 papers - 33 comparisons), Engineering (3391 papers - 345 comparisons), Life Sciences (4142 papers - 195 comparisons), Physical Sciences & Mathematics (15694 papers - 731 comparisons), and Social and Behavioral Sciences (844 papers - 170 comparisons). Below this is a 'Comparisons' section with tabs for 'Comparisons', 'Papers', 'Visualizations', 'Reviews', and 'Lists'. The 'Comparisons' tab is active, showing a list of comparison entries. The first entry is 'Systematic Literature Review (SLR) Tools analysed based on General Features' in the 'Information Science' field, with 16 contributions and 0 visualizations, dated 04-03-2024. The second entry is 'R0 estimates for infectious diseases' in the 'Virology' field, with 6 contributions and 0 visualizations, dated 04-04-2024. The third entry is 'Machine learning of pre-harvesting crop/fruit parameters to minimize overall losses in farming production' in the 'Plant Cultivation, Pl...' field. To the right of the comparison list, there are three additional widgets: 'Load Toots' (with a note about Mastodon widget loading), 'ORKG stories' (with a link to 'Find out more'), and 'Join ORKG!' (with a 'Sign up' button). At the bottom right, there is a 'Knowledge base for science' section with a book icon and the text: 'We are flooded with new publications in research every day and it is increasingly challenging to keep up'.

The ORKG revolutionizes how **scientific knowledge** is communicated and processed, making it **actionable** for machines and **researchers** to navigate, compare, and review **vast amounts of information** efficiently.

# Research Knowledge Graphs

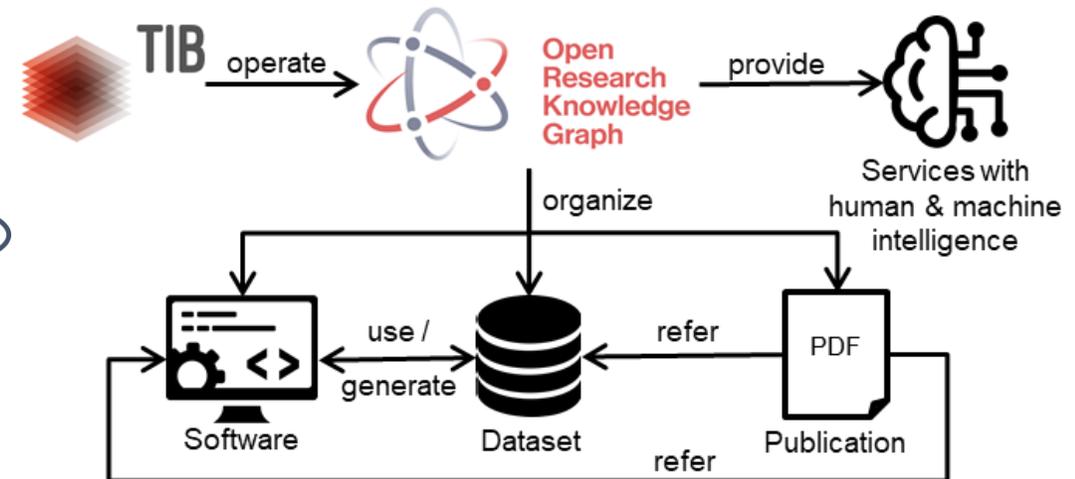
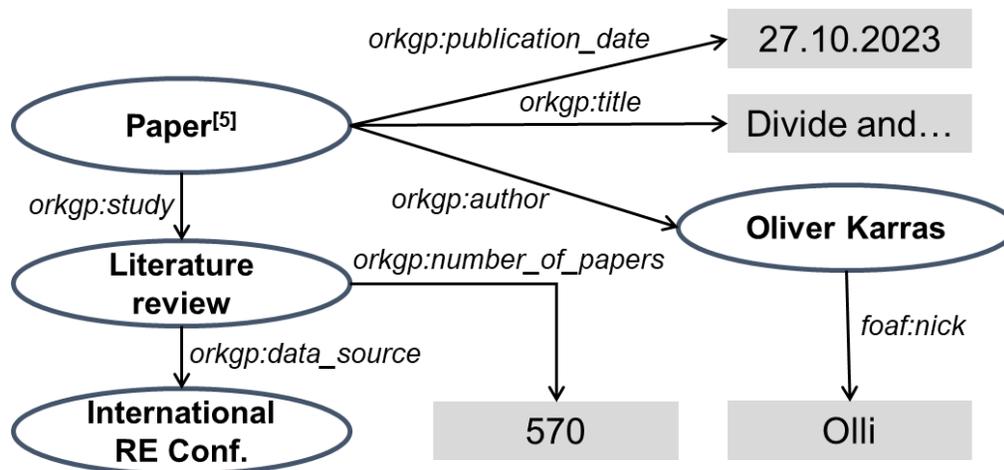
**Research Knowledge Graphs** (RKGs) are a **technology** for organizing scientific (meta-)data in a

- **Flexible, fine-grained, and semantic** representation
- That is **understandable** and **processable** by humans and machines

# Research Knowledge Graphs

**Research Knowledge Graphs (RKGs)** are a **technology** for organizing scientific (meta-)data in a

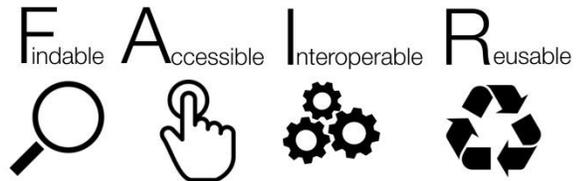
- **Flexible, fine-grained, and semantic** representation
- That is **understandable** and **processable** by humans and machines



The **ORKG** is a **ready-to-use** and **sustainably** operated **infrastructure** with **services** that uses a **cross-discipline RKG** for the **long-term** and **openly available organization** of scientific (meta-)data according to the **FAIR** data principles.

[5] Karras et al.: *Divide and Conquer the EmpiRE: A Community-Maintainable Knowledge Graph of Empirical Research in Requirements Engineering*. 2023 ACM/IEEE International Symposium on Empirical Software Engineering and Measurement (ESEM), DOI: [10.1109/ESEM56168.2023.10304795](https://doi.org/10.1109/ESEM56168.2023.10304795), 2023.

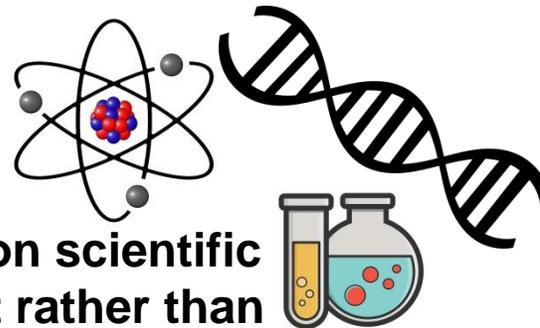
# ORKG Objectives



**Make research FAIR**



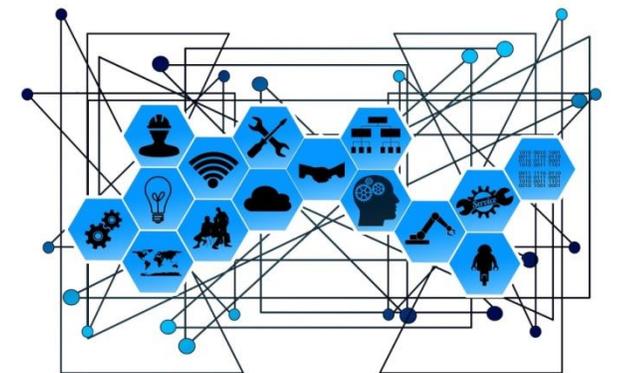
**Provide long-term  
available and up-to-date  
overviews**



**Focus on scientific  
content rather than  
pure metadata**



**Foster collaboration**



**Tackle interdisciplinary  
challenges**

# Structured Representation of Scientific Knowledge

## Mining User Requirements from Application Store Reviews Using Frame Semantics

Nishant Jha and Anas Mahmoud<sup>(✉)</sup>

The Division of Computer Science and Engineering, Louisiana State University,  
Baton Rouge, LA 70803, USA  
njha1@lsu.edu, mahmoud@csc.lsu.edu

**Abstract.** *Context and motivation:* Research on mining user reviews in mobile application (app) stores has noticeably advanced in the past few years. The majority of the proposed techniques rely on classifying the textual description of user reviews into different categories of technically informative user requirements and uninformative feedback. *Question/Problem:* Relying on the textual attributes of reviews often produces high dimensional models. This increases the complexity of the classifier and can lead to overfitting problems. *Principal ideas/results:* We propose a novel semantic approach for app review classification. The proposed approach is based on the notion of semantic role labeling, or characterizing the lexical meaning of text in terms of semantic frames. Semantic frames help to generalize from text (individual words) to more abstract scenarios (contexts). This reduces the dimensionality of the data and enhances the predictive capabilities of the classifier. Three datasets of user reviews are used to conduct our experimental analysis. Results show that semantic frames can be used to generate lower dimensional and more accurate models in comparison to text classification methods. *Contribution:* A novel semantic approach for extracting user requirements from app reviews. The proposed approach enables a more efficient classification process and reduces the chance of overfitting.

**Keywords:** Requirements elicitation · Application stores · Classification

### 1 Introduction

Mobile application markets, or app stores (e.g., Google Play and Apple App Store), represent a unique model of service-oriented business. Such platforms have created an unprecedented opportunity for app developers to directly monitor the opinions of a large population of end-users of their software [25]. Through app stores feedback services, app users can directly share their experience in the form of textual reviews and meta-data (e.g., star ratings). Analyzing large datasets of app store reviews has revealed that they contain substantial amounts of up-to-date technical information. Such information can be leveraged by app developers to help them maintain and sustain their apps in a highly-competitive

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P. Grünbacher and A. Perini (Eds.): REFSQ 2017, LNCS 10153, pp. 273–287, 2017.  
DOI: 10.1007/978-3-319-54045-0\_20



## Mining User Requirements from Application Store Reviews Using Frame Semantics

📅 2017    📁 Software Engineering    👤 Nishant Jha    👤 Anas Mahmoud

Published in: *Requirements Engineering: Foundation for Software Quality*

DOI: 10.1007/978-3-319-54045-0\_20

### User Feedback Classification (Template)

Add to comparison

#### Contribution data

← Back    User has experiment → Machine Learning Experiment ↻

Classification category	Feature request
Has dataset	seel.cse.lsu.edu/data/refsq17.zip
Has feature	Bag of words
Has method	Support vector machines
Has result	P = 0.45, R = 0.68, F1 = 0.54, etc.

Provenance    Timeline

**Added on**  
12 May 2021

**Added by**  
 Oliver Karras

**Contributors**  
Kheir Eddine  
Eduard C. Groen  
Oliver Karras

 Assign to observatory



# Using FAIR Scientific Knowledge

How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved?

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Nishant Jha and Anas Mahmoud<sup>(✉)</sup>

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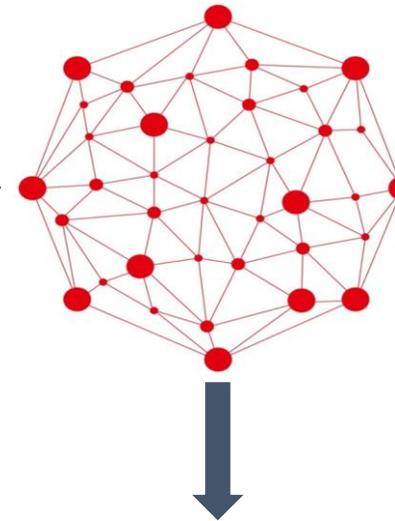
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DOI: 10.1007/978-3-319-54045-0\_20

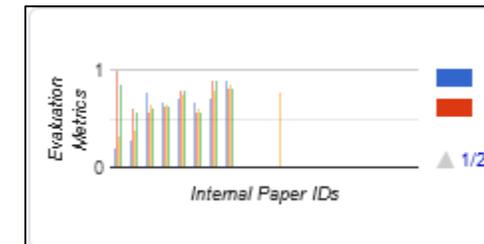


Natural  
language  
question



Answer

Properties	Software Feature Request Detection in Issue Tracking Systems <small>2016 - User Feedback Classification</small>	Mining User Requirements from Application Store Reviews Using Frame Semantics <small>2017 - User Feedback Classification</small>
Has dataset <span style="float: right;">▼</span>	<a href="https://zenodo.org/record/56907#.YKT_NudCRPY">https://zenodo.org/record/56907#.YKT_NudCRPY</a>	<a href="https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip">https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip</a>
	<a href="https://sites.google.com">https://sites.google.com</a>	



# ORKG Comparisons

Acknowledgement of creators

DOI

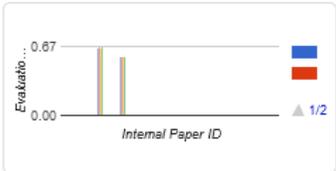
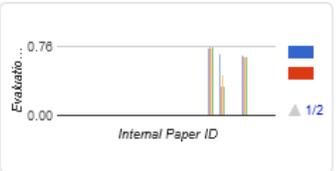
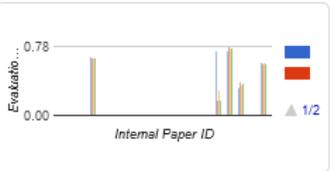
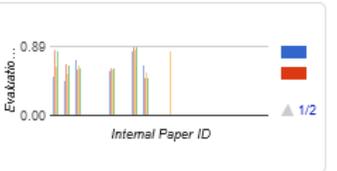
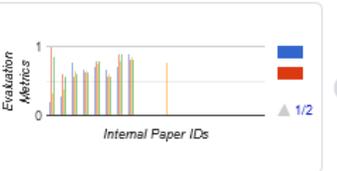
Visualizations

Interactive filtering

## Overview of Approaches that Classify User Feedback as Feature Request ☆ 👁

📅 June 2021   👤 Oliver Karras   👤 Eduard C. Groen

This overview shows the classification results of approaches that use the machine learning algorithms Naïve Bayes, Support Vector Machines, and Decision Trees C4.5 in combination with the machine learning features Bag of Words or Term Frequency - Inverse Document Frequency to classify user feedback as feature request.

Properties	Software Feature Request Detection in Issue Tracking Systems <i>User Feedback Classification - 2016</i>	Mining User Requirements from Application Store Reviews Using Frame Semantics <i>User Feedback Classification - 2017</i>	Mining Twitter Feeds for Software User Requirements <i>User Feedback Classification - 2017</i>	Automatic Classification of Non-Functional Requirements from Augmented App User Reviews <i>User Feedback Classification - 2017</i>	Bug reports simply classified <i>User Feedback Classification - 2017</i>
has dataset	<a href="https://zenodo.org/record/56907#.YKT_NudCRPY">https://zenodo.org/record/56907#.YKT_NudCRPY</a>	<a href="https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip">https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip</a> <a href="https://sites.google.com/site/appsuserreviews/">https://sites.google.com/site/appsuserreviews/</a> <a href="https://seel.cse.lsu.edu/data/refsq17.zip">seel.cse.lsu.edu/data/refsq17.zip</a>	<a href="https://seel.cse.lsu.edu/data/re17.zip">seel.cse.lsu.edu/data/re17.zip</a>	Not available	<a href="https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip">https://mast.informatik.uni-hamburg.de/wp-content/uploads/2014/03/REJ_data.zip</a>

# ORKG Comparisons are Citable

## Comparison of Studies on Germany's Energy Supply in 2050 ★ 🔍

📅 November 2021
👤 Felix Kullmann
👤 Jan Göpfert
👤 Oliver Karras
👤 Patrick Kuckertz
👤 Sören Auer
👤 Markus Stocker

👤 Peter Markewitz
👤 Leander Kotzur
👤 Detlef Stolten



Share  
🌐  
🐦  
🌐  
🔗

This comparison compares the results of various studies analyzing a future low-carbon energy system for Germany. The focus of this study comparison is electricity generation. In the future, however, other essential characteristics of the respective energy system designs in the individual studies will be listed. Installed capacity is given in GW and electricity generation is given in TWh.

DOI: <https://doi.org/10.21203/rs.3.rs-11111111/v1>



Oliver Karras

Bearbeiten
Löschen

### Comparison of studies on Germany's energy supply in 2050

Autoren Felix Kullmann, Peter Markewitz, Detlef Stolten, Oliver Karras, Patrick Kuckertz, Leander Kotzur, Jan-Maris Göpfert, Sören Auer, Markus Stocker

Publikationsdatum 2021

Ausgabe FZJ-2022-00782

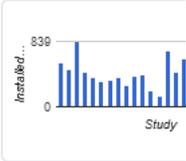
Verlag Technoökonomische Systemanalyse

Beschreibung This comparison compiles the results from various studies analyzing a future low-carbon energy system for Germany. The focus of this study comparison is electricity generation. In the future, however, other essential characteristics of the respective energy system designs in the individual studies will be listed. Installed capacity is given in GW and electricity generation is given in TWh.

Zitate insgesamt Zitiert von: 2



Google Scholar-Artikel [Comparison of studies on Germany's energy supply in 2050](#)  
F Kullmann, P Markewitz, D Stolten, O Karras... - 2021  
Zitiert von: 2 [Ähnliche Artikel](#)



Properties

has energy sources ▾

Den Weg zu einem treibhausgasneutralen Deutschland ressourcenschonend gestalten

Contribution 1 - 2019

- [all sources](#)
- [bioenergy](#)
- [geothermics](#)
- [hydropower](#)
- [import](#)
- [net import](#)

net import
net import
net import
net import

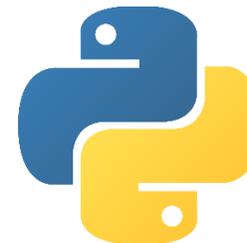
# So far so Good, but...

## ...what can we do with machine-actionable scientific knowledge?

Anything we want!

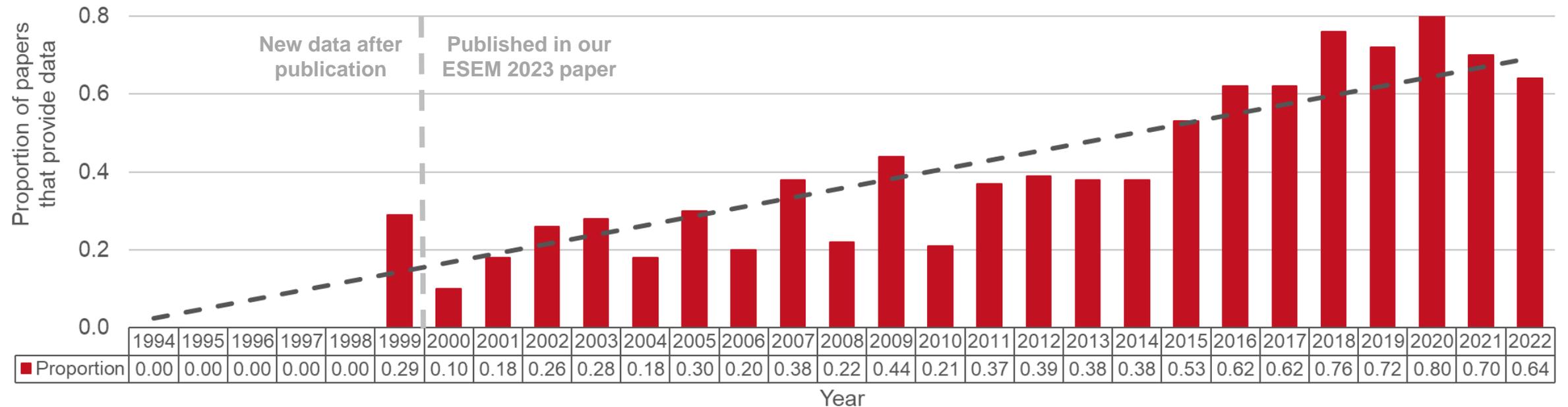
1. Papers, comparisons, and visualizations in the ORKG are **openly available for (re-)use and extension to anyone.**
2. The ORKG provides several **interfaces** for processing the data, e.g., to develop novel search, retrieval, mining, and assistance applications.

{ REST : API }



# (Re-)use and Extension of Scientific Knowledge

**How has the provision of data (materials used, raw data collected, and study results identified) in RE evolved? [5, 6, 7]**  
(Based on 680 papers in the ORKG)



[5] Karras et al.: *Divide and Conquer the EmpiRE: A Community-Maintainable Knowledge Graph of Empirical Research in Requirements Engineering*. 2023 ACM/IEEE International Symposium on Empirical Software Engineering and Measurement (ESEM), DOI: [10.1109/ESEM56168.2023.10304795](https://doi.org/10.1109/ESEM56168.2023.10304795), 2023.

[6] Project on GitHub: <https://github.com/okarras/EmpiRE-Analysis>

[7] Interactive Jupyter notebook: <https://mybinder.org/v2/gh/okarras/EmpiRE-Analysis/HEAD?labpath=%2Fempire-analysis.ipynb>

# ORKG Observatory

Empirical Software Engineering | Observatory Edit

This observatory works to provide a community-maintainable knowledge graph of empirical research in software engineering. Our goal is to continuously acquire and curate comprehensive knowledge about empirical research applied in scientific publications in the research field of software engineering and its subfields, such as requirements engineering. In this way, we want to provide a comprehensive, up-to-date, and long-term overview of the state-of-the-art on empirical research in software engineering.

We are currently working on a knowledge graph of empirical research in requirements engineering.

For this purpose, we have developed a corresponding ORKG template (cf. <https://orkg.org/template/R186491>).

Contact: [oliver.karras@tib.eu](mailto:oliver.karras@tib.eu)

Research problems

- empirical research in software engineering
- empirical research in requirements engineering

Organizations



Members

 **Oliver Karras**  
TIB - Leibniz Information Centre for Science and Technology

 **Jil Kluender**  
Leibniz University Hannover

Content | 688 items

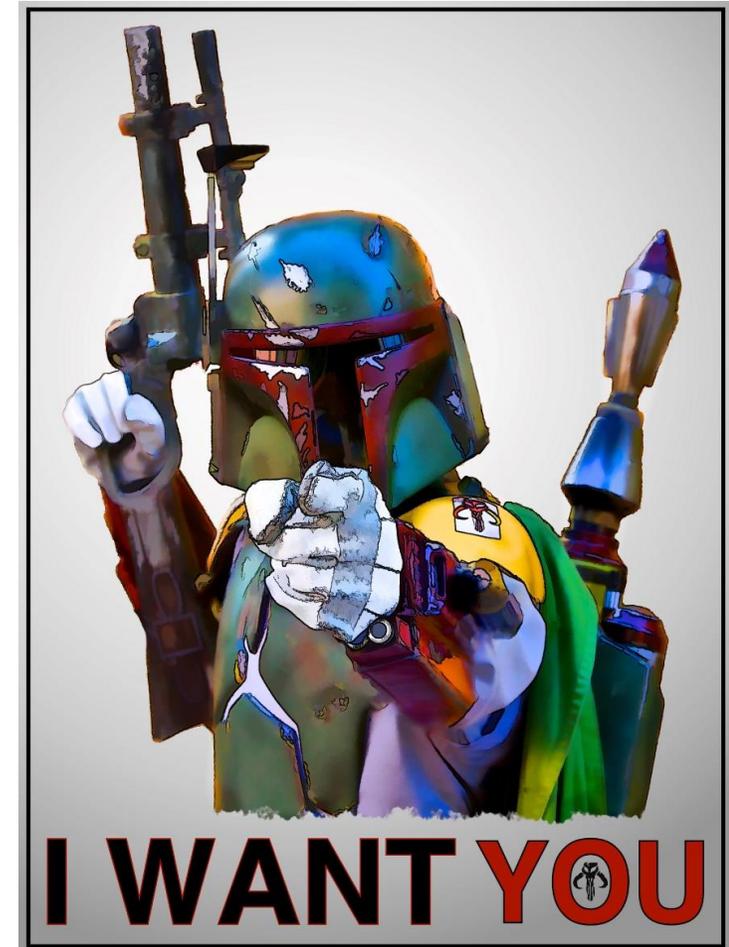
Show:  Paper  Comparison  Visualization Top recent

- **Open** groups maintain topics in the ORKG
- **Central** access point for the community to all curated contents

# Who organizes this scientific knowledge in the ORKG?

# Who Organizes this Scientific Knowledge in the ORKG?

The heart of the ORKG is its **CROWD**:  
**Researchers** from any discipline!



# How can We Contribute to the ORKG?

## 1. While writing a paper:

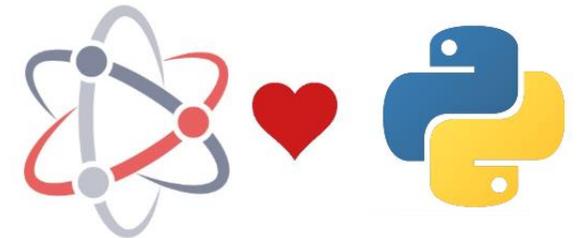
- SciKGTex: LaTeX package for FAIR annotations in a paper (embedded in the PDF) that can be imported into ORKG
- <https://github.com/Christof93/SciKGTex>



SciKGTex

## 2. While developing an analysis:

- Python package (& R package in development) for FAIR annotations in analysis scripts that can be imported into the ORKG
- <https://pypi.org/project/orkg/>



## 3. At any time:

- Manually using the ORKG Frontend to describe papers
- (Semi-)automatically using the ORKG REST API



Open  
Research  
Knowledge  
Graph

{ REST : API }

# What We Will Learn to Contribute to the ORKG

## 1. While writing a paper:

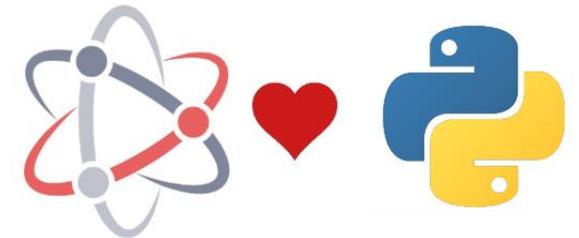
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SciKGTex

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Open  
Research  
Knowledge  
Graph

{ REST : API }

# Conclusion

**Mining User Requirements from Application Store Reviews Using Frame Semantics**

Nishaant Jha and Anas Mahmoud<sup>✉</sup>

The Division of Computer Science and Engineering, Louisiana State University,  
Baton Rouge, LA 70803, USA  
njha@lsu.edu, mahmoud@ce.lsu.edu

**Abstract.** Context and motivation: Research on mining user reviews in mobile application (app) stores has noticeably advanced in the past few years. The majority of the proposed techniques rely on classifying the textual description of user reviews into different categories of technically informative user requirements and uninformative feedback. **Question/Problem:** Relying on the textual attributes of reviews often produces high dimensional models. This increases the complexity of the classifier and can lead to overfitting problems. **Principal idea/solution:** We propose a novel semantic approach for app review classification. The proposed approach is based on the notion of semantic role labeling, or characterizing the lexical meaning of text in terms of semantic frames. Semantic frames help to generalize from text (individual words) to more abstract scenarios (contexts). This reduces the dimensionality of the data and enhances the predictive capabilities of the classifier. Three datasets of user reviews are used to conduct our experimental analysis. Results show that semantic frames can be used to generate lower dimensional and more accurate models in comparison to text classification methods. **Contribution:** A novel semantic approach for extracting user requirements from app reviews. The proposed approach enables a more efficient classification process and reduces the chance of overfitting.

**Keywords:** Requirements elicitation · Application stores · Classification

**1 Introduction**

Mobile application markets, or app stores (e.g., Google Play and Apple App Store), represent a unique model of service-oriented business. Such platforms have created an unprecedented opportunity for app developers to directly monitor the opinions of a large population of end-users of their software [2]. Through app stores feedback services, app users can directly share their experience in the form of textual reviews and meta-data (e.g., star ratings). Analyzing large datasets of app store reviews has revealed that they contain substantial amounts of up-to-date technical information. Such information can be leveraged by app developers to help them maintain and sustain their apps in a highly-competitive

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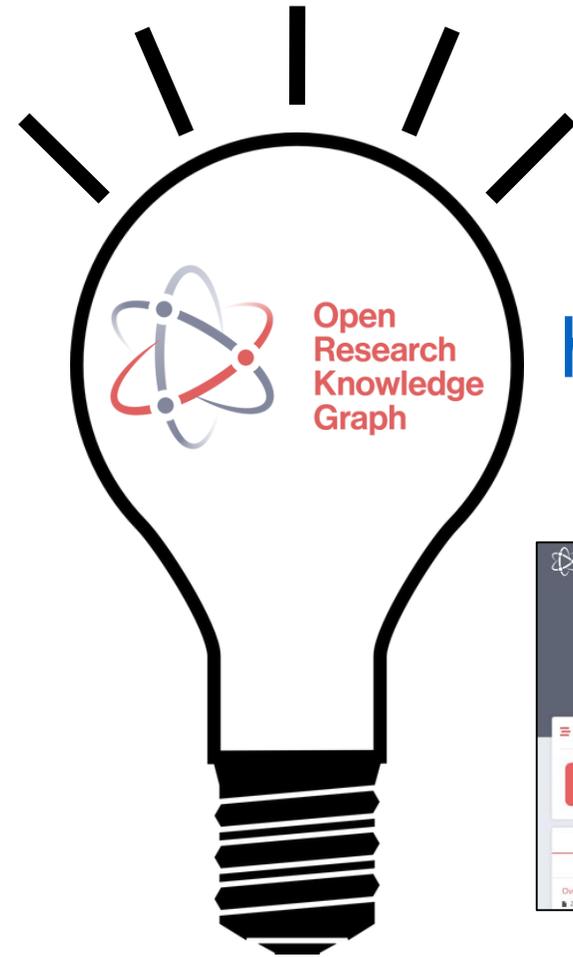
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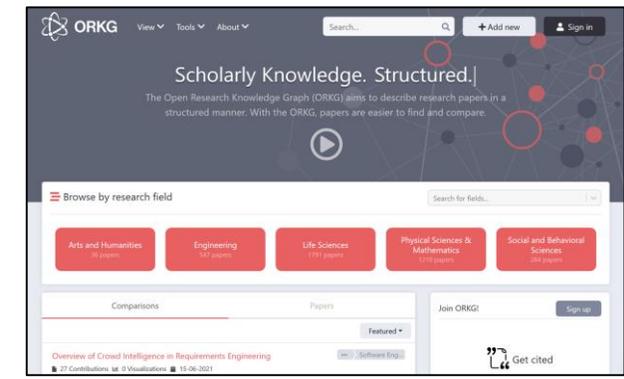
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<https://orkg.org>



**Let's bring scholarly communication and open science in Requirements Engineering to the 21<sup>st</sup> century!**

# Further Reading



## Open Access ORKG Book

- Celebrating the 5th anniversary of ORKG
- A practical guide for new and advanced users
  - ORKG's terms and concepts
  - ORKG's approach
  - ORKG's technology
  - ORKG's success stories

<https://cuvillier.de/de/shop/publications/9037-open-research-knowledge-graph>

